

Uncertainty Quantification of GNSS RO Retrieval Products

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Status of CLARREO-related tasks from past year

1. RO bending angle climatology for high altitude initialization (Nov 2015)
 - Algorithm being implemented/tested at JPL processing system
2. Trends and variability of the tropical width from RO, reanalysis, and climate models (May 2016)
 - Paper in preparation
 - To be presented at AMS Annual Meeting 2017
3. Collaboration with Knuteson & Feltz (UW) on understanding biases between IR & RO strato temperatures.
4. obs4MIPs monthly-averaged RO temperature and geopotential height datasets have been released.

Motivation: Establishing GNSS RO as reference observations

- Following the GRUAN (GCOS Reference Upper Air Network) paradigm:
 - ✓ *Is traceable to an SI unit or an accepted standard*
 - ✓ ***Provides a comprehensive uncertainty analysis***
 - ✓ *Is documented in accessible literature*
 - ✓ *Is validated (e.g. by intercomparison or redundant observations)*
 - ✓ *Includes complete meta data description*
 - ✓ ***Important to distinguish contributions from systematic error and random error***

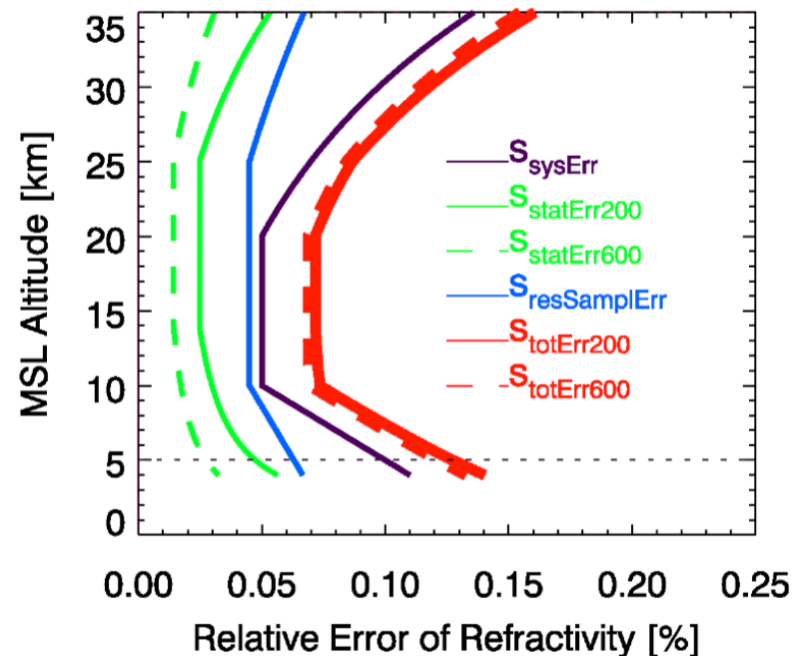
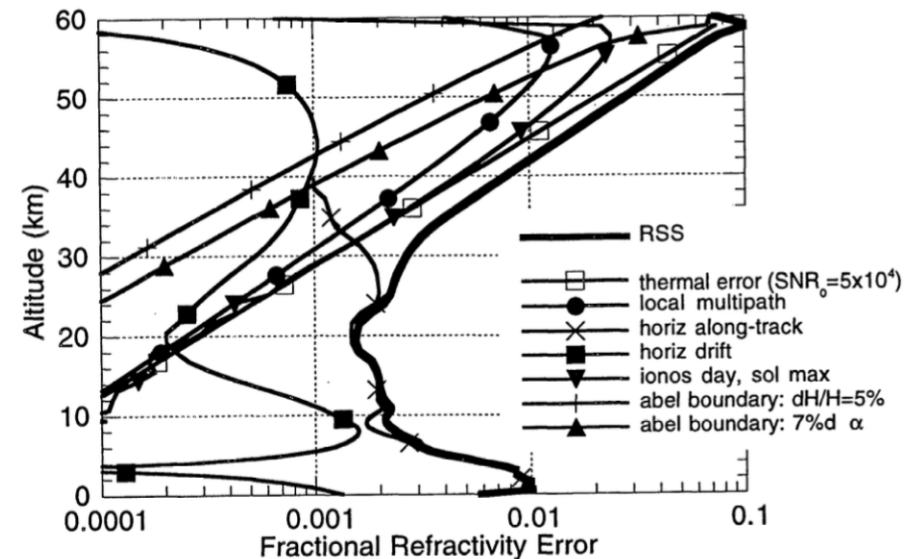
Existing works

Kursinski et al., JGR, 1997

- Comprehensive analysis of multiple error sources.
- Did not separate systematic and random errors
- Derived under limited sets of parameters.

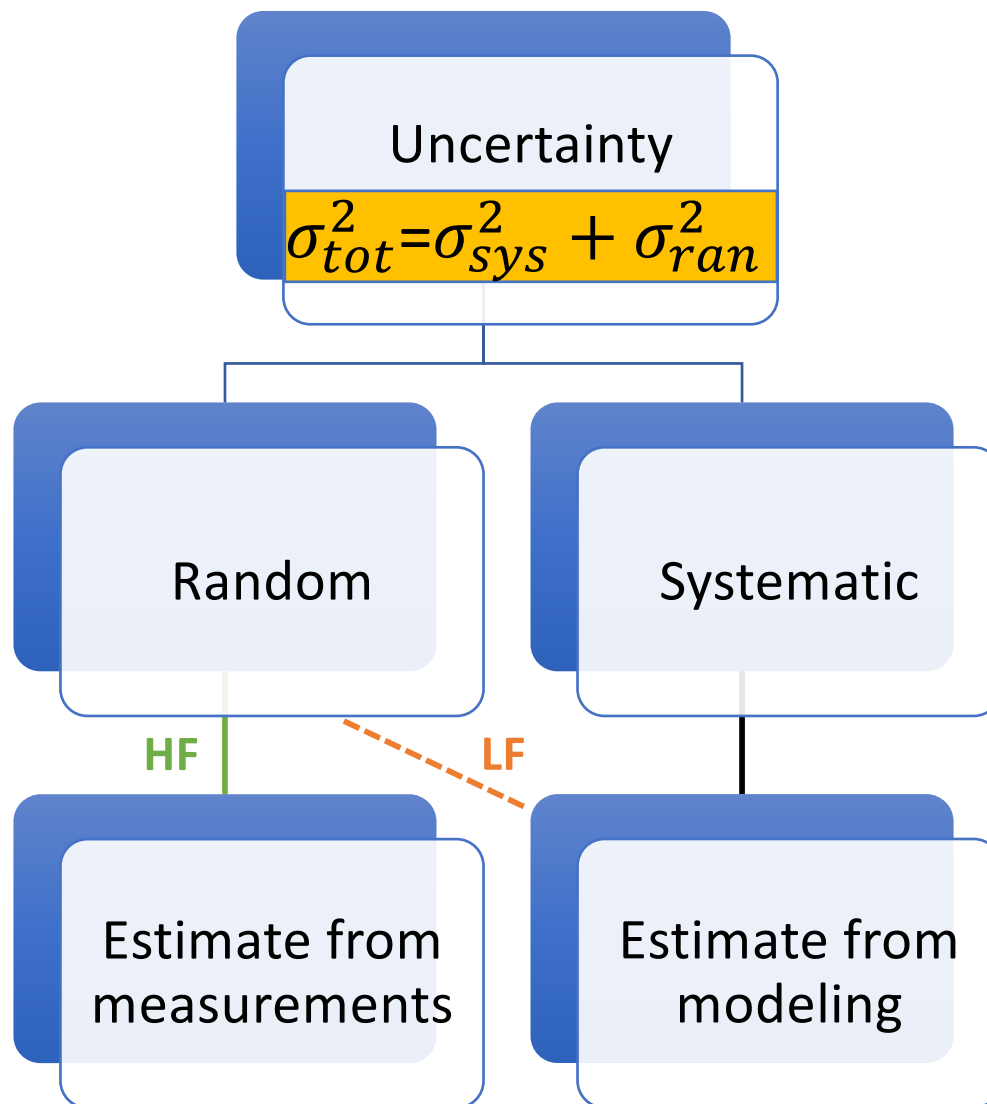
Scherlin-Pirscher et al., AMT, 2011

- Separated systematic and random errors (plus sampling error estimates).
- Considered only climatological averages.



Independent uncertainty estimates specific to a retrieval system are desirable.

Approach to per datum uncertainty estimation



Random = noise that averages down as $\sim 1/\sqrt{N}$, where N is the number of observations.

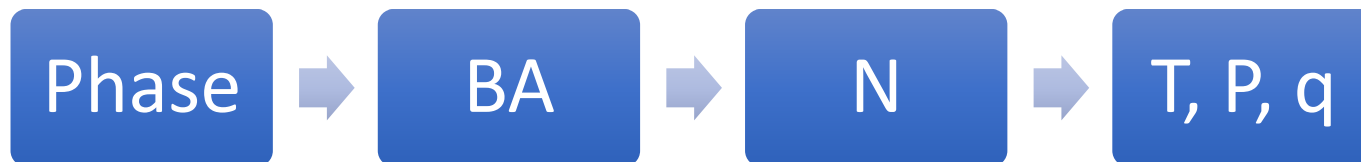
HF = High-frequency variations (uncorrelated noise from sample to sample)

LF = Low-frequency variations (from multipaths and possibly orbit/clock)

Estimating random errors

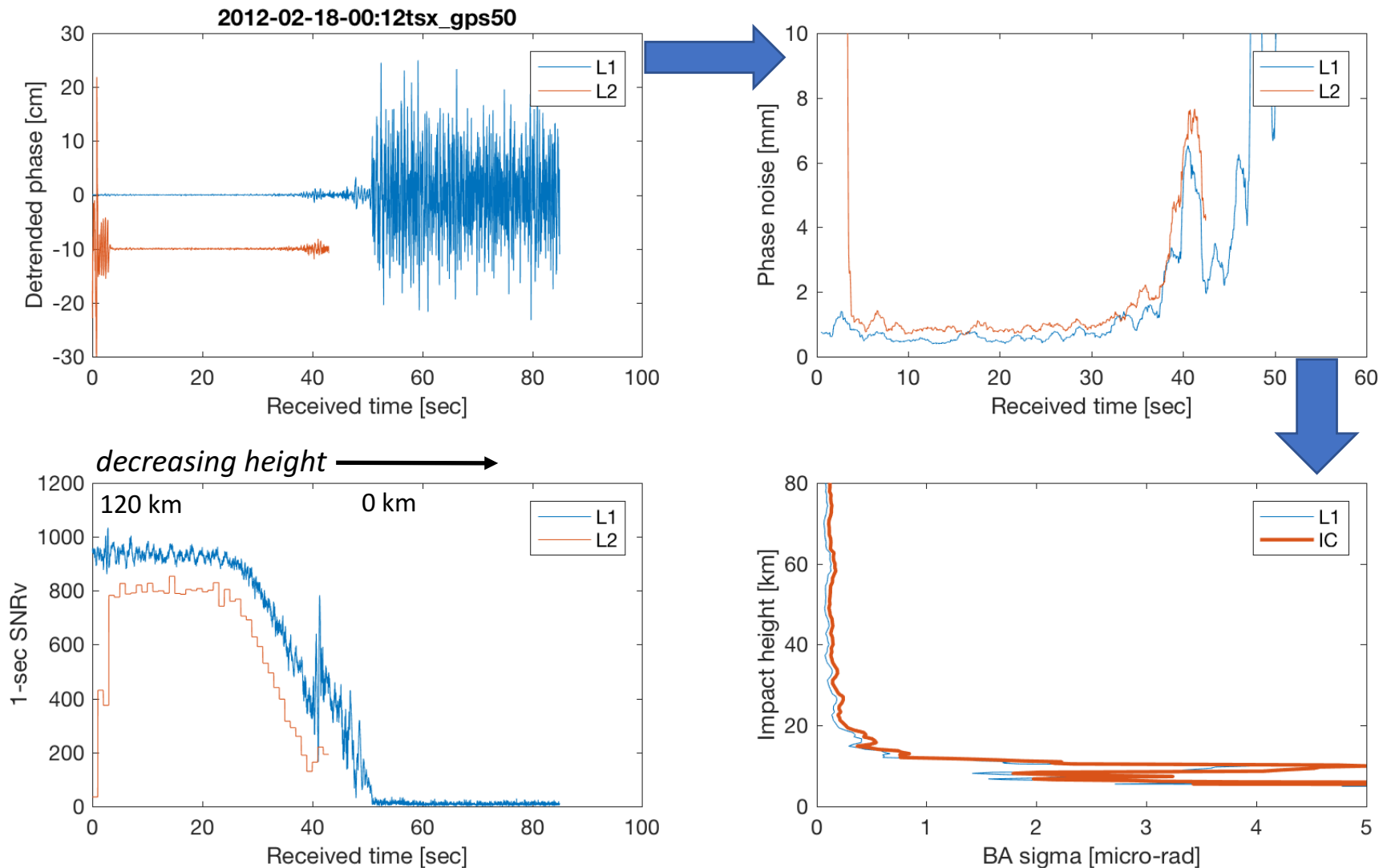
Estimate phase noise from the excess phase data:

- Detrend phase and compute standard deviation over an interval containing many data points (e.g., 1 sec = 50 pts)
- Propagate phase uncertainty through retrieval chain (*analytically* or *Monte Carlo*)



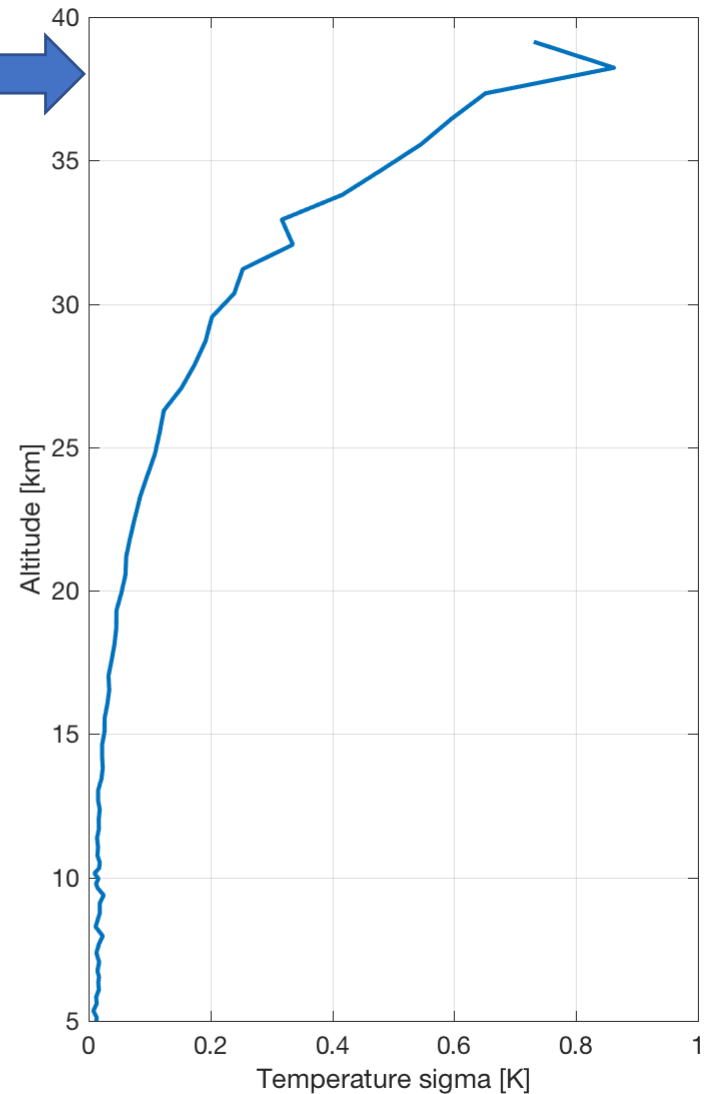
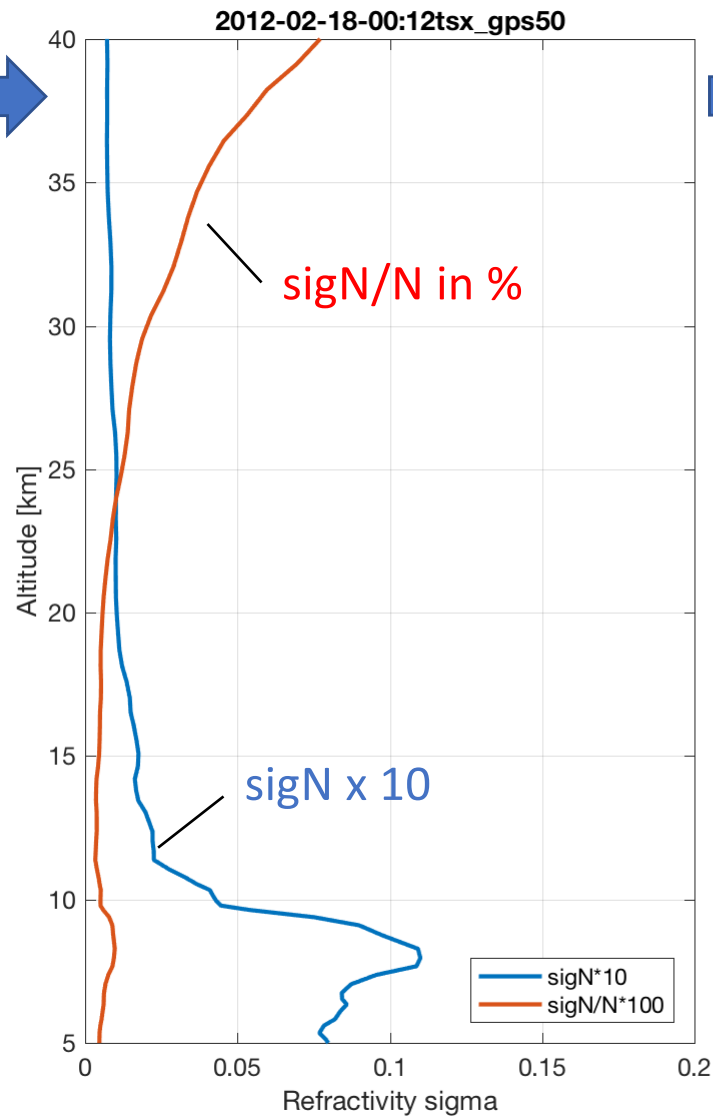
This method accounts only for “high-frequency” noise. It will not capture “low-frequency” phase variations such as those from local multipath.

Example 1: Phase to BA (Random)

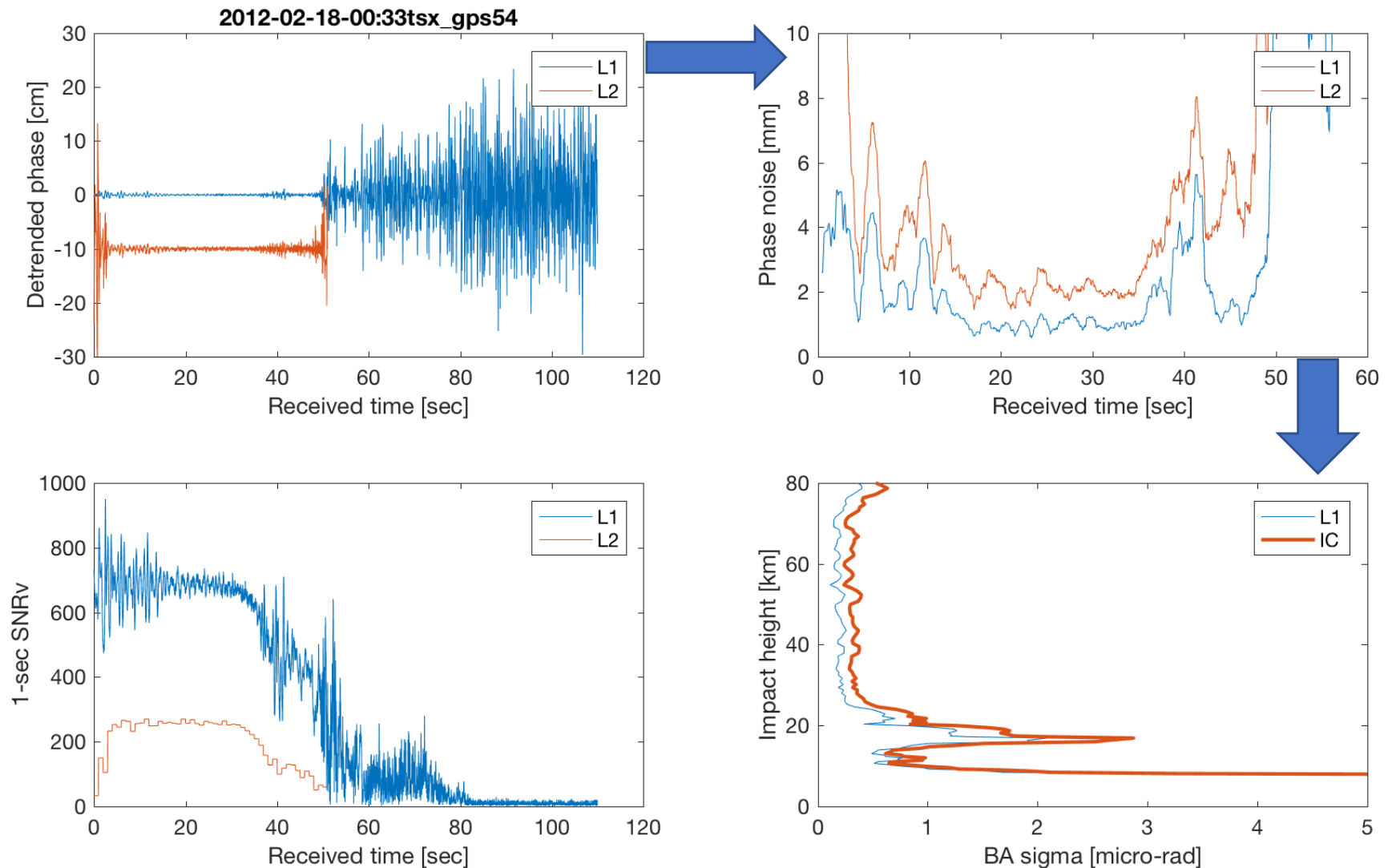


Example 1: BA to N to T (Random)

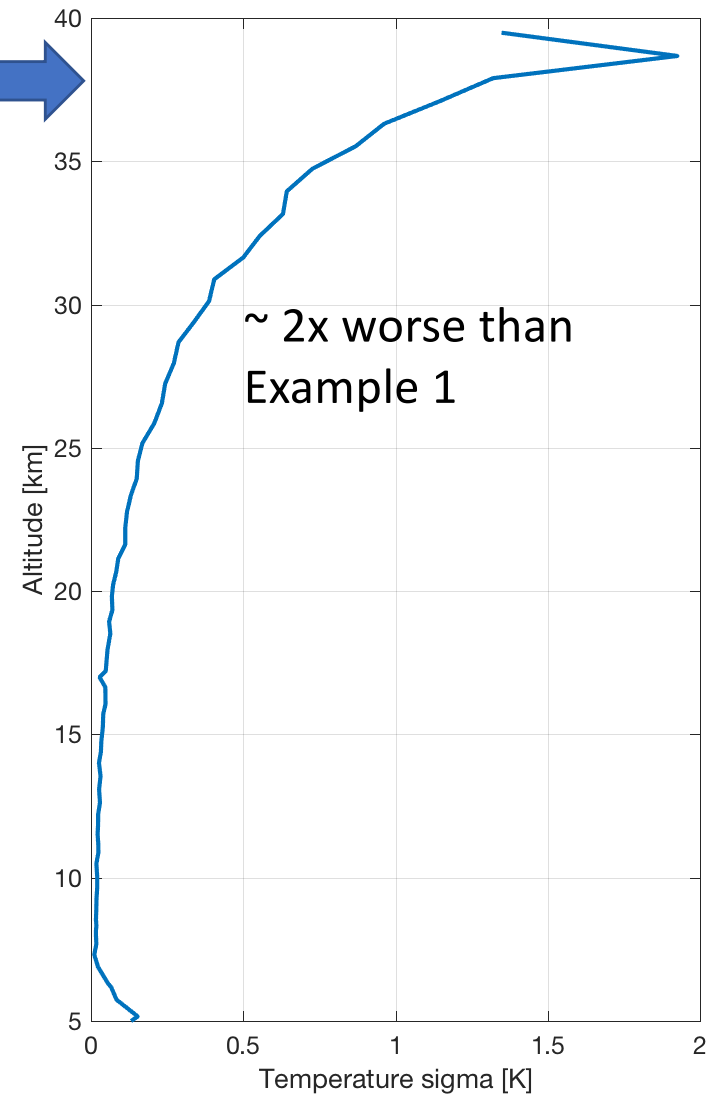
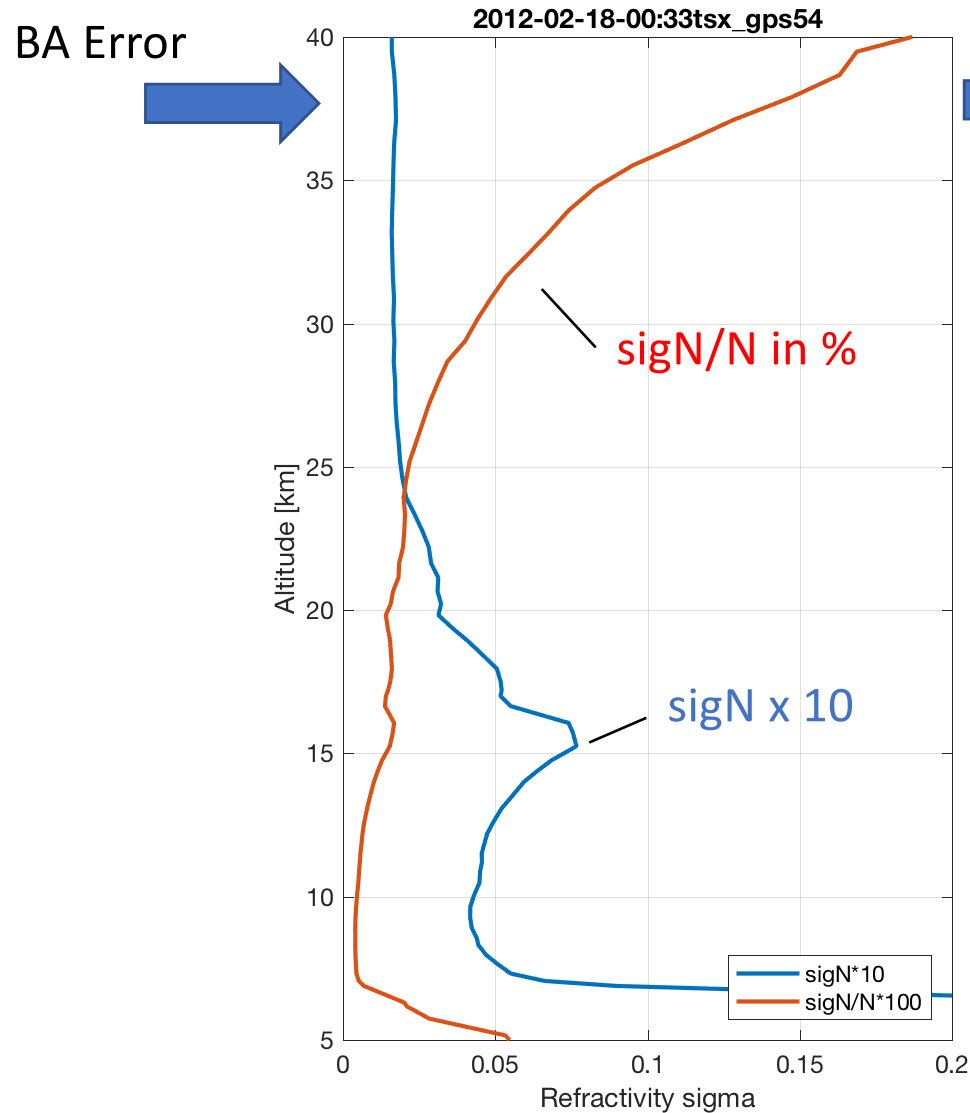
BA Error



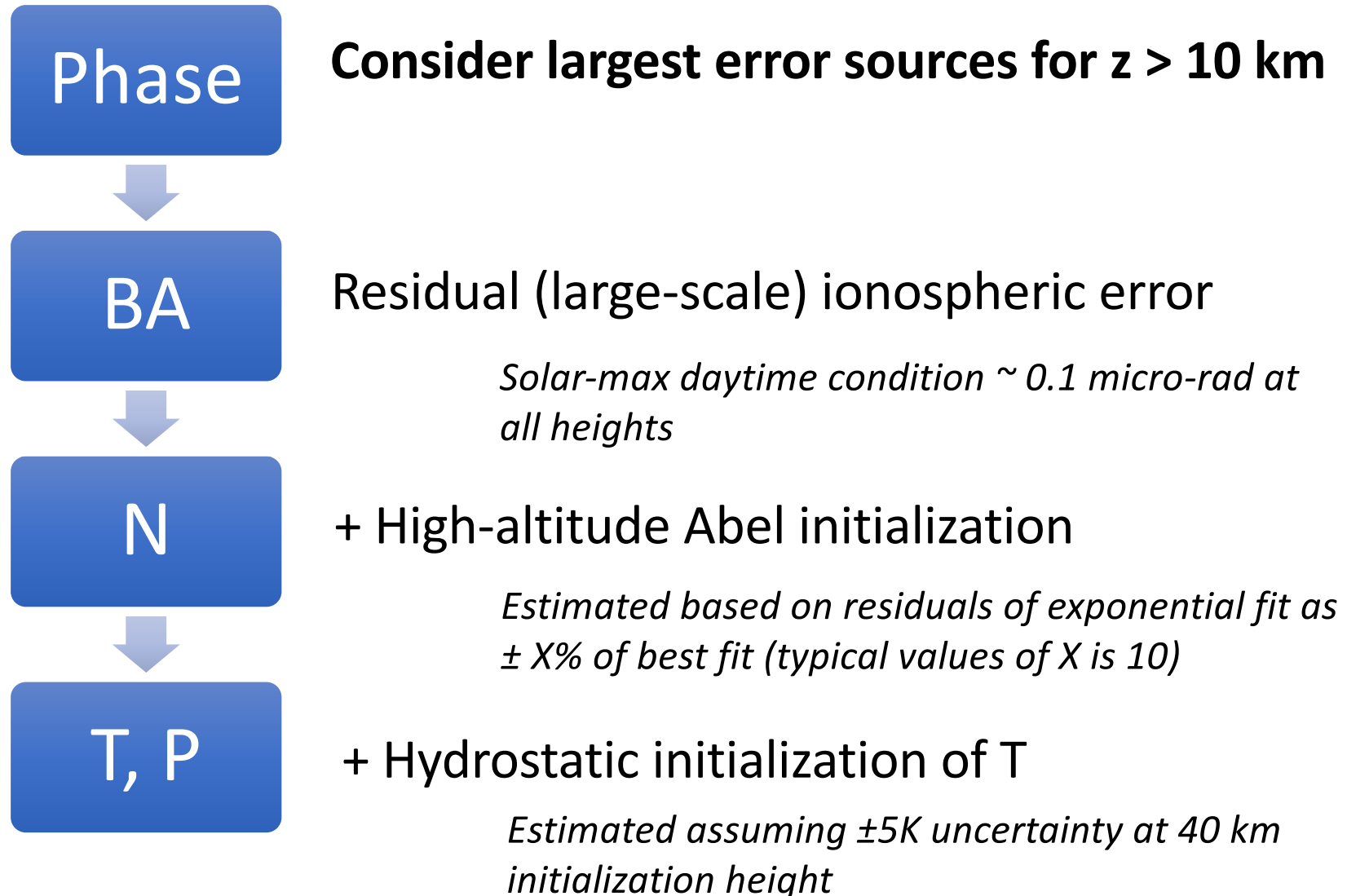
Example 2: Phase to BA (Random)



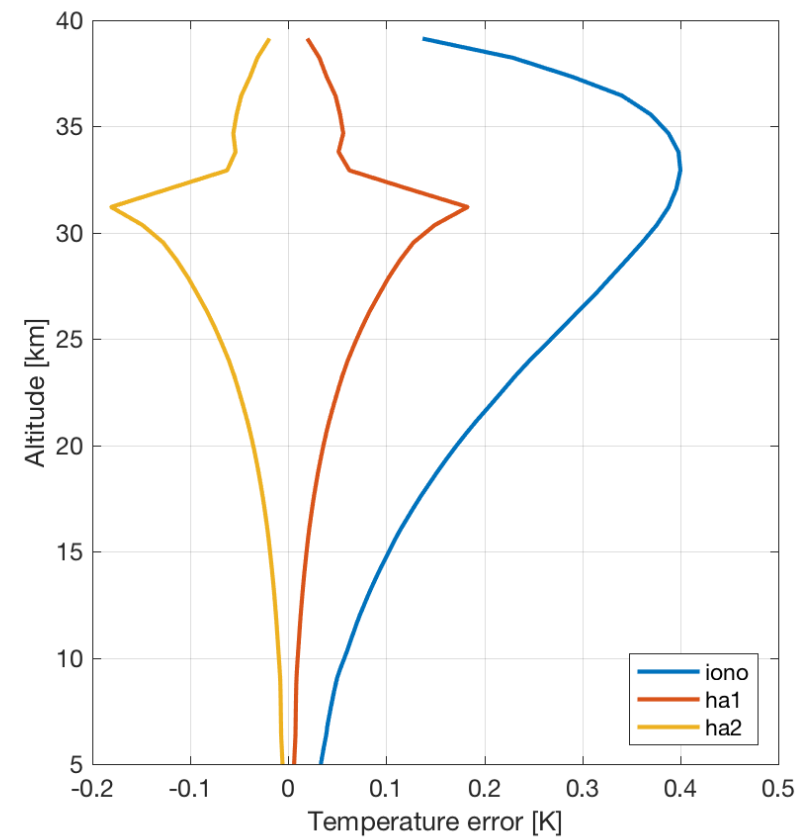
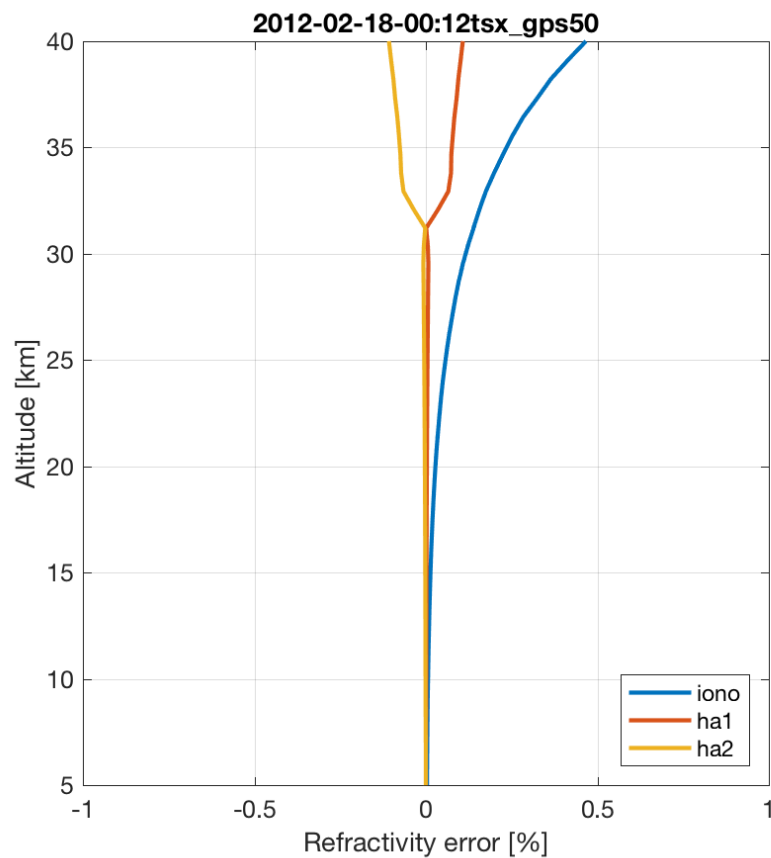
Example 2: BA to N to T (Random)



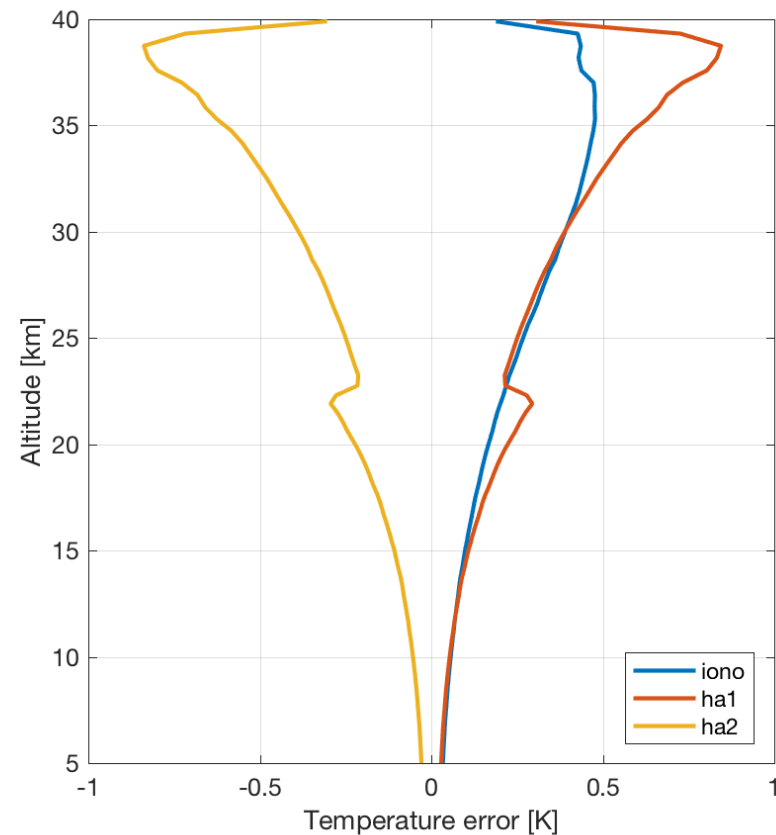
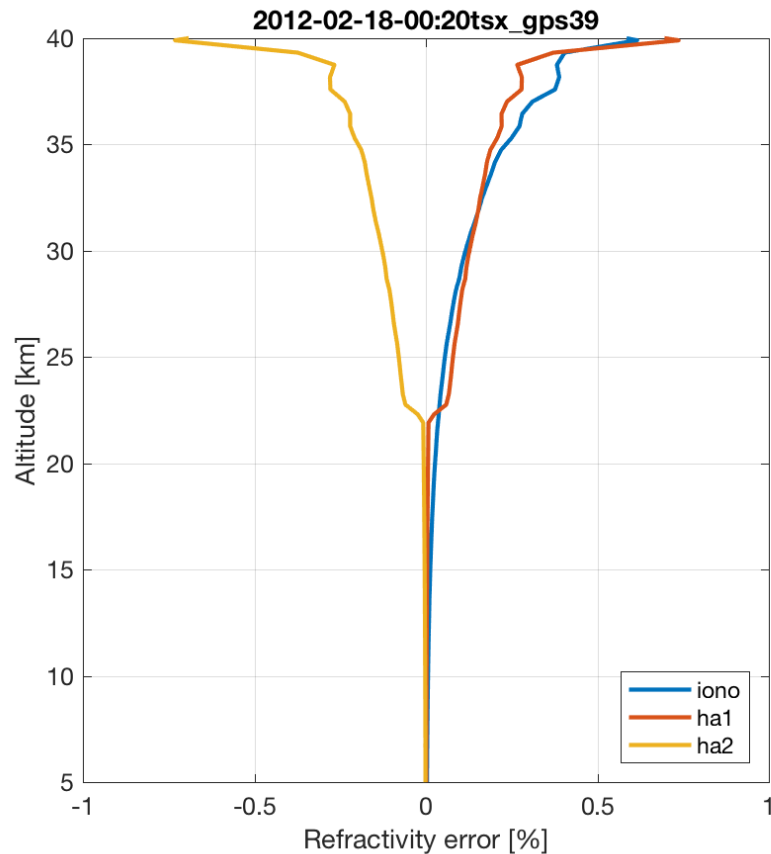
Estimating systematic errors



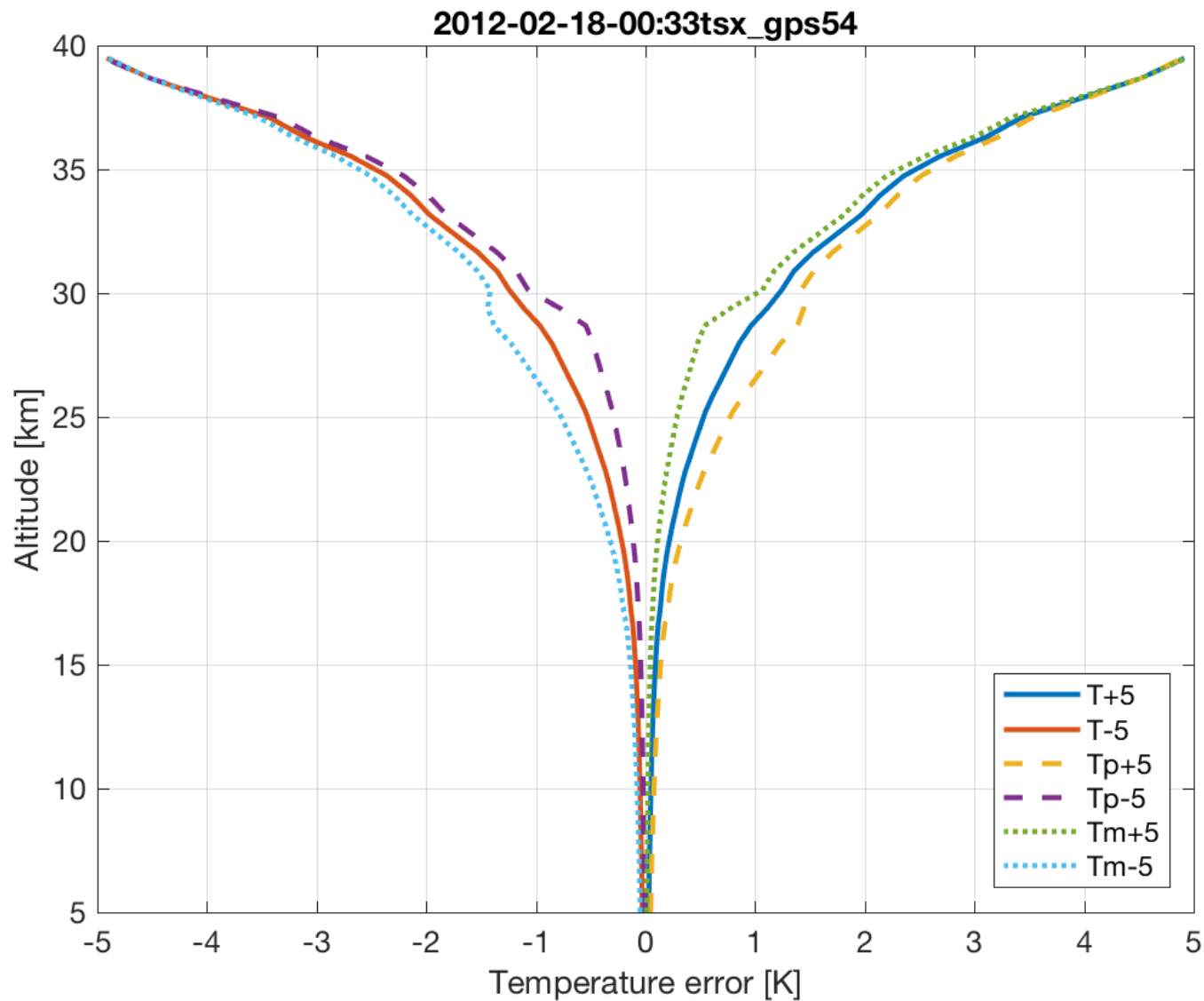
Example 1: Systematic, iono+ha



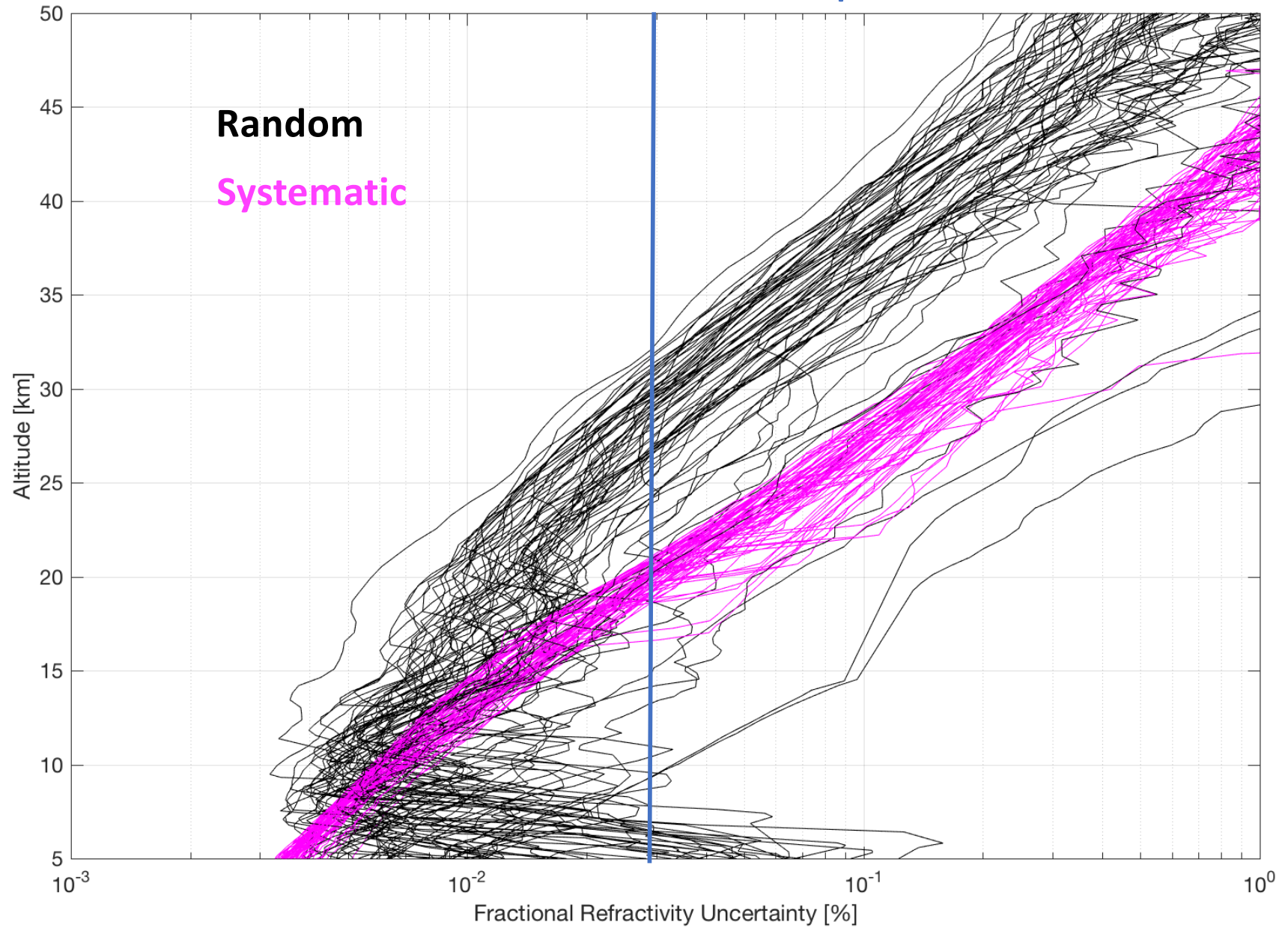
Example 2: Systematic, iono+ha



Example 2: Systematic, T init



0.03% CLARREO requirement



Summary

- Progress towards per datum uncertainty characterization of RO retrieval products.
- Dominant error sources have been taken into account.
- Systematic iono error can be better estimated (currently only an upper bound).
- Lower troposphere error characterization needs additional work (modeling tracking error & noise propagation thru nonlinear retrieval).
- Uncertainty estimates need to be verified.
- Retrieval errors from non-spherical atmosphere were not considered here (but were in Kursinski et al. 1997); they can be interpreted as *representation errors*.